

# Lecture 7

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2017-02-28

# Tidying data with the tidyverse

```
library(tidyverse) # loads dplyr, ggplot2, tidyr, etc
```

- ▶ Reading: Hadley Wickham's article on tidy data, available from GitHub (Notes folder) or canvas.
  - ▶ Today's notes closely follow the `tidyr` vignette.
- ▶ “Tidy” data is ready for analysis, with one row for each sampled unit and columns for the variables measured on the units.
  - ▶ Often classify variables as “explanatory” or “response”
- ▶ Tabular data and repeated measures data are often not in tidy form.
- ▶ Examples:
  - ▶ Tabular data on new tuberculosis cases from WHO
  - ▶ Repeated measures from Billboard Top 100

# Tuberculosis (TB) cases

```
tb <- read.csv("tb.csv",stringsAsFactors=FALSE)
dim(tb)
```

```
## [1] 5769    23
```

```
tb[1:4,1:6]
```

```
##   iso2 year new_sp new_sp_m04 new_sp_m514 new_sp_m014
## 1   AD 1989     NA          NA          NA          NA
## 2   AD 1990     NA          NA          NA          NA
## 3   AD 1991     NA          NA          NA          NA
## 4   AD 1992     NA          NA          NA          NA
```

```
names(tb)[1:20]
```

```
## [1] "iso2"          "year"          "new_sp"        "new_sp_m04"
## [5] "new_sp_m514"   "new_sp_m014"   "new_sp_m1524"  "new_sp_m2534"
## [9] "new_sp_m3544"  "new_sp_m4554"  "new_sp_m5564"  "new_sp_m65"
## [13] "new_sp_mu"     "new_sp_f04"    "new_sp_f514"   "new_sp_f014"
## [17] "new_sp_f1524"  "new_sp_f2534"  "new_sp_f3544"  "new_sp_f4554"
```

# Structure of TB table

- ▶ First column is 2-letter country code, second is year, third is number of new cases for that country/year.
- ▶ Then come TB counts for different gender/age categories.
  - ▶ `new_sp` is “new cases by positive pulmonary smear assay”
  - ▶ gender is `m` or `f`
  - ▶ two special age categories 0-4, 5-14,
  - ▶ age categories 0-14, 15-24, 25-34, 35-44, 45-54, 55-65, 65+, unknown (`u`)
- ▶ Gender/age columns are not variables, they are data on the observed units.
- ▶ Tidy data would have one row for each country, year, gender and age category, with a column of counts

# Billboard Top 100 rankings of songs

```
bb <- read.csv("billboard.csv", stringsAsFactors = FALSE)
dim(bb)
```

```
## [1] 317 83
```

```
bb[1:3, 1:6]
```

```
##   year artist.inverted      track time genre date.entered
## 1 2000 Destiny's Child Independent Women Part I 3:38  Rock    2000-09-23
## 2 2000          Santana             Maria, Maria 4:18  Rock    2000-02-12
## 3 2000   Savage Garden      I Knew I Loved You 4:07  Rock    1999-10-23
```

```
names(bb)[c(1:10, ncol(bb))]
```

```
## [1] "year"          "artist.inverted" "track"
## [4] "time"          "genre"           "date.entered"
## [7] "date.peaked"   "x1st.week"       "x2nd.week"
## [10] "x3rd.week"     "x76th.week"
```

## Structure of the Billboard table

- ▶ Columns `year` through `date.peaked` describe the song, then `x1st.week` through `x76th.week` are the chart positions for the first through 76th weeks.
  - ▶ If a song is on the chart for less than 76 weeks, its position is NA for any missing weeks.
- ▶ Weeks are not variables, they are the time data for the time series.

# Tidying the Billboard data

- ▶ Main step is to consolidate, or “gather” ranking in the different weeks into a `rank` variable.
- ▶ Before gathering, will select/rename some of the variables.
- ▶ After gathering, will create some new variables and sort the data frame.

# Select and rename

- ▶ Won't need time or genre.
  - ▶ `select()` from `dplyr` can use `-` to de-select
- ▶ Rename `artist.inverted`
  - ▶ `rename()` from `dplyr` takes arguments of the form `newname = oldname`

```
bb %>% select(-time,-genre) %>%  
  rename(artist = artist.inverted) -> bb
```



## Gather the weeks into a “long” version of the Billboard data

- ▶ Leave each song info variable as-is.
- ▶ The data, or “values”, are the chart positions.
- ▶ The weeks are descriptors or “keys” for these values.
- ▶ We want to create key-value pairs for each observation.
  - ▶ There will be missing values, which we can remove.
- ▶ The `gather()` function from `tidyr` gathers specified columns into keys (e.g., `week`) and values (e.g., `rank`).

## gather() for the Billboard data

```
bblong <- gather(bb, week, rank, x1st.week:x76th.week, na.rm=TRUE)
head(bblong, n=4)
```

##	year	artist	track	date.entered	date.peaked
## 1	2000	Destiny's Child	Independent Women Part I	2000-09-23	2000-11-18
## 2	2000	Santana	Maria, Maria	2000-02-12	2000-04-08
## 3	2000	Savage Garden	I Knew I Loved You	1999-10-23	2000-01-29
## 4	2000	Madonna	Music	2000-08-12	2000-09-16

  

##	week	rank
## 1	x1st.week	78
## 2	x1st.week	15
## 3	x1st.week	71
## 4	x1st.week	41

## More cleaning suggested in the vignette

- ▶ Extract week numbers from week variable
- ▶ Coerce date.entered to a Date object
- ▶ Calculate the date of each ranking based on the date it entered the charts and the week.
- ▶ Sort (“arrange”) on artist, track and week.

```
bblong %>% mutate(week = parse_number(week),  
                  date = as.Date(date.entered) + 7*(week-1)) %>%  
  select(-date.entered) %>% # don't need date.entered anymore  
  arrange(artist, track, week) -> bb  
head(bb, n=3)
```

```
##      year artist                                     track date.peaked week rank  
## 1 2000    2 Pac Baby Don't Cry (Keep Ya Head Up II) 2000-03-11     1    87  
## 2 2000    2 Pac Baby Don't Cry (Keep Ya Head Up II) 2000-03-11     2    82  
## 3 2000    2 Pac Baby Don't Cry (Keep Ya Head Up II) 2000-03-11     3    72  
##           date  
## 1 2000-02-26  
## 2 2000-03-04  
## 3 2000-03-11
```

## Tidying the TB data

- ▶ Recall structure of the data: country, year, count of new cases, counts of new cases by gender/age categories.

```
names(tb)[1:10]
```

```
## [1] "iso2"          "year"          "new_sp"        "new_s  
## [5] "new_sp_m514"  "new_sp_m014"  "new_sp_m1524" "new_s  
## [9] "new_sp_m3544" "new_sp_m4554"
```

- ▶ Main step is to “gather” TB prevalence in the different gender/age categories into a count variable.
  - ▶ Complicated by the coding of gender/age categories
- ▶ Before gathering, will remove unneeded variables and add country names to supplement 2-letter codes.

# Remove variables

- ▶ Won't need overall count
- ▶ Special categories 0-4 and 5-14 overlap with 0-14 so remove
- ▶ Age unknown not useful for analysing trends, so remove

```
tb <- select(tb, -new_sp, -contains("04"), -contains("514"),  
             -new_sp_mu, -new_sp_fu)  
tb[1:3, 1:10]
```

```
##   iso2 year new_sp_m014 new_sp_m1524 new_sp_m2534 new_sp_m3544  
## 1   AD 1989           NA             NA             NA             NA  
## 2   AD 1990           NA             NA             NA             NA  
## 3   AD 1991           NA             NA             NA             NA  
##   new_sp_m4554 new_sp_m5564 new_sp_m65 new_sp_f014  
## 1           NA           NA           NA           NA  
## 2           NA           NA           NA           NA  
## 3           NA           NA           NA           NA
```

## Add country names to supplement country codes

- ▶ I found a translation of the ISO-2 country codes at [<http://data.okfn.org/data/core/country-list>] and saved as `countryCodes.csv` in the Notes folder.
  - ▶ Exercise: Find out which ISO-2 codes are in `tb` but not in `countryCodes.csv`, google the missing codes, and add the country names to `cc` manually.

```
cc <- read.csv("countryCodes.csv", stringsAsFactors = FALSE)
# cc has columns "Name" and "Code". "Code" matches "iso2" in tb
```

- ▶ Exercise: Find out which ISO-2 codes are in `tb` but not in `countryCodes.csv`, google the missing codes, and add the country names to `cc` manually.

```
cc <- rbind(cc, data.frame(Name=c("Netherlands Antilles", "Yugoslavia"),
                           Code=c("AN", "YU")))
tb <- inner_join(cc, tb, by = c("Code" = "iso2"))
tb[1:2, 1:6]
```

##	Name	Code	year	new_sp_m014	new_sp_m1524	new_sp_m2534
## 1	Afghanistan	AF	1980	NA	NA	NA
## 2	Afghanistan	AF	1981	NA	NA	NA

## Gather counts for demographic groups

- Create demographic variable `demog` and count variable `count` by gathering over all variables except `Name`, `Code` and `year`.

```
tblong <- gather(tb,demog,count,-Name,-Code,-year,na.rm=TRUE)
head(tblong)
```

##	Name	Code	year	demog	count
## 13	Afghanistan	AF	1997	new_sp_m014	0
## 14	Afghanistan	AF	1998	new_sp_m014	30
## 15	Afghanistan	AF	1999	new_sp_m014	8
## 16	Afghanistan	AF	2000	new_sp_m014	52
## 17	Afghanistan	AF	2001	new_sp_m014	129
## 18	Afghanistan	AF	2002	new_sp_m014	90

## Separate gender from age category.

- First remove new\_sp\_, then separate result on first column (help(separate))

```
maxlen <- max(nchar(tblong$demog))
tblong %>% mutate(demog = substr(demog,8,maxlen)) %>%
  separate(demog, into=c("gender","agecat"),sep=1) -> tb
head(tb)
```

##	Name	Code	year	gender	agecat	count
## 1	Afghanistan	AF	1997	m	014	0
## 2	Afghanistan	AF	1998	m	014	30
## 3	Afghanistan	AF	1999	m	014	8
## 4	Afghanistan	AF	2000	m	014	52
## 5	Afghanistan	AF	2001	m	014	129
## 6	Afghanistan	AF	2002	m	014	90

```
save(tb,file="tb.RData")
```